A Combined Communication and Synchronization Optimization Algorithm for One-Way Communication

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Abstract

As against the conventional send/receive model, the one-way communication is a new paradigm which decouples message transmission and synchronization. This opens up new opportunities not only to further optimize communication but also to minimize synchronization costs. In this paper, we present a generalized technique which makes use of global dataflow framework to optimize communication and synchronization in the context of one way communication model. The uniqueness of our approach stems from the fact that it can take all types of HPF-like alignments and distributions into account and is more powerful than any existing algorithm for eliminating redundant synchronization messages. Preliminary results with several example programs demonstrate that our approach is successful in minimizing the number of data and synchronization messages.

Keywords: one-way communication, synchronization, compiler, data-parallel languages, optimization.

1 Introduction

Most of the current compilers for distributed-memory architectures rely on `send` and `recv` primitives to implement communication. The impact of this approach is twofold. First, this technique combines synchronization with communication; that is, the data messages also carry implicit synchronization information. While this relieves the compiler from inserting synchronization messages to maintain data integrity, in some programs, separating synchronization messages from data messages may actually improve the performance. Second, the compiler has the job of matching `send` and `recv` operations in order to guarantee correct execution.

In this paper, we focus on compilation of programs annotated by HPF-like directives with one-way communication operations `Put` and `Synch`. As an example, consider Figure 1(a). In this figure, a consumer sends a `Synch` message to the producer informing that the producer can *put* data in a buffer physically located in the consumer's memory. After receiving the `Synch` message, the producer deposits the data in that buffer. We note that the `Synch` operation is necessary for the repetition of this communication; that is, when the producer wants to deposit new data into the buffer, it should know that the consumer processor had consumed the old data in the buffer.

After briefly discussing the fundamental concepts used in this paper in Section 2, in Section 3 we show how the communication sets as well as the producers and the consumers manipulated by the `Put` operation are implemented on top of the existing `send`/`recv` type of communication framework in the PARADIGM compiler [3]. Having determined those, the next issue is to minimize the number of `Put` communications as well as the communication volume. Section 4 presents an algorithm to achieve this goal. Our algorithm can take arbitrary control flow into account and can optimize programs with all types of HPF-like alignments and distributions, including block-cyclic distributions. To our knowledge, it is the first algorithm that can optimize communication globally in the existence of block-cyclic distributions. It is based on a linear algebra framework introduced by Ancourt *et.al.* [2] and is quite general in the sense that by redefinition of a predicate several previous approaches can be simulated.

Clearly, in a compilation framework based on the `Put` operation, the correct ordering of memory accesses has to be imposed by the compiler using the synchronization primitives. A straightforward approach inserts a `Synch` operation just before each `Put` operation as shown in Figure 1(a). The next question to be addressed then is whether or not every `Synch` operation inserted that way is always necessary. The answer is "no", and Section 5 proposes an algorithm to eliminate redundant synchronization messages. We refer to a `Synch` operation as "redundant" if its functionality can be fulfilled by other data communications or other `Synch` operations occurring in the program. The basic idea is to use another message in the reverse direction between the same pair of processors in place of the `Synch` call as shown in Figure 1(b). In that case, we say that the communication $\mathtt{t}_j$ kills the synchronization requirement of communication `i`. We show that our algorithm is very fast and more powerful than the previous work in synchronization elimination. This is because (1) it is very accurate in eliminating the redundant synchronization, as using the Omega library [11, 14], it works on processor pair granularity; (2) it can eliminate a synchronization message by using several data messages; (3) it handles block, cyclic, and block-cyclic distributions in a unified manner, whereas the previous approaches either work on virtual processor grids only or use an extension of regular section descriptors which are inherently inaccurate; and (4) it is preceded by a global communication optimization algorithm which itself eliminates a lot of synchronization messages. To show the idea behind the algorithm, consider Figure 2(a), where eight processors are involved in a communication. The arrows indicate the direction of communication. Figure 2(b) shows the `Synch` messages required for the repetitions of this communication. Figures 2(c) and 2(d), on the other hand, depict two other communications that occur between the repetitions of the communication given in Figure 2(a). Our synchronization elimination algorithm can detect that the communications given in Figures 2(c) and 2(d), *together*, kill the synchronization requirement of the first communication.

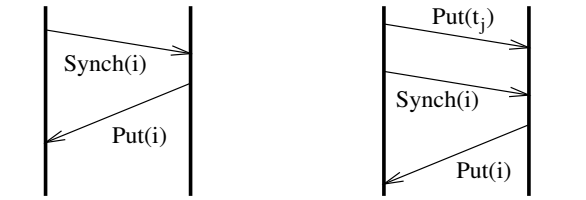
In Section 6, we give preliminary results on several benchmark programs. We believe that these are also the first results from a comprehensive evaluation of synchronization elimination in one-way communication. We discuss the related work in Section 7 and conclude the paper with a summary in Section 8.

2 Basic Concepts

In this section we briefly describe the concepts used in this paper. We focus on structured programs with conditional statements and nested loops but without arbitrary goto statements. A *basic block* is a sequence of consecutive statements in which the flow of control enters at the beginning and leaves at the end without the possibility of branching except perhaps at the end [1]. A *control flow graph* (CFG) is a directed graph constructed by basic blocks and represents the flow-of-control information of the program. For our purposes, the CFG can be thought of as a directed graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ where each $v \in \mathcal{V}$ represents either a basic block or a (reduced) interval which represents a loop, and each $e \in \mathcal{E}$ represents an edge between blocks. In this paper, depending on the context, we use the term *node* interchangeably for a statement, a block or an interval. Two unique nodes s and t denote the start and terminal nodes respectively of a CFG. One might think of these nodes as dummy statements. We define the sets of all successors and predecessors of a node n as $succ(n) = \{m \mid (n, m) \in \mathcal{E}\}$ and $pred(n) = \{m \mid (m, n) \in \mathcal{E}\}$, respectively. We say node i *dominates* node j in the CFG (written as $j \in dom(i)$), if every path from s to j goes through i. We assume that prior to communication analysis, any edge that goes directly from a node with more than one successor to a node with more than one predecessor is split by introducing a dummy node. Our technique for minimizing the communication volume and the number of messages is based on *interval analysis* [1]. Interval analysis consists of a *contraction* phase and an *elimination* phase. For programs written in a structured language, an interval corresponds to a loop. The contraction phase collects information about what is generated and what is killed inside each interval. Then the interval is reduced to a single node and annotated with the information collected. This is a recursive procedure and stops when the reduced CFG contains no more cycles. In each step of the expansion phase, on the other hand, a node (reduced interval) is expanded, and the information regarding the nodes in that interval is computed.

3 Producers and Consumers

We assume that all loop bounds and subscript expressions are affine functions of enclosing loop indices or constants. Under this condition, a loop nest, an array and a processor grid can all be represented as bounded polyhedra. The PARADIGM compiler currently uses the owner-computes rule [9, 20], which assigns each computation to the processor that owns the data being computed.



(a) (b)

Figure 1: (a) one-way communication with `Put` operation. (b) elimination of a `Synch` message.

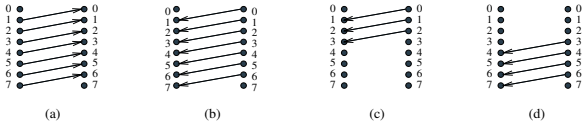


Figure 2: Communication and synchronization messages for the first loop of the program in Figure 6(a).

```
   REAL X(x_l:x_u), Y(y_l:y_u)
!HPF$ TEMPLATE T(t_l:t_u)
!HPF$ PROCESSORS PROC(p_l:p_u)
!HPF$ ALIGN X(j) WITH T(α*j+β)
!HPF$ DISTRIBUTE T(CYCLIC(C)) ON TO PROC

  DO t = 1, TIMES /* time-step loop */
    ...
    < synchronization for Y using Synch operations >
S:  < communication for Y using Put operations >
    DO i = i_l, i_u
      ...
      X(γ_L*i+θ_L) = ··· Y(γ_R*i+θ_R) ···
      ...
    END DO
    ...
  END DO
```

(a)

SET	PROJECTION FUNCTION
Producers(S)	$[p', q', d] \mapsto [q']$
Consumers(S)	$[p', q', d] \mapsto [p']$
ProducersFor(S, p)	$[p, q', d] \mapsto [q']$
ConsumersFor(S, q)	$[p', q, d] \mapsto [p']$
PutSet(S, p, q)	$[p, q, d] \mapsto [d]$
SendSet(S)	$[p', q', d] \mapsto [q', p']$
Pending(S)	$[p', q', d] \mapsto [p', q']$

(b)

Figure 3: (a) Generic loop. (b) Projection functions to manipulate sets (p and q are symbolic names).

Consider the one-dimensional generic loop i shown in Figure 3(a). Let $\mathcal{R}_{\mathcal{L}}(i)$ = X(γ_L*i+θ_L) and $\mathcal{R}_{\mathcal{R}}(i)$ = X(γ_R*i+θ_R). Assuming symbols p and q denote two processors, we define the following sets where S is the communication statement and $\vee$ and $\wedge$ are logical "or" and "and" operations respectively:

$$
\begin{aligned}
\texttt{Own}(\texttt{X}, q) &= \{d \mid d \in \texttt{X} \wedge \text{ is owned by } q\} \\
\texttt{Producers}(\texttt{S}) &= \{q' \mid \exists i, p' \text{ such that } \mathcal{R}_{\mathcal{R}}(i) \in \texttt{Own}(\texttt{Y}, q') \wedge \mathcal{R}_{\mathcal{L}}(i) \in \texttt{Own}(\texttt{X}, p') \wedge i_l \leq i \leq i_u \wedge q' \neq p'\} \\
\texttt{Consumers}(\texttt{S}) &= \{p' \mid \exists i, q' \text{ such that } \mathcal{R}_{\mathcal{R}}(i) \in \texttt{Own}(\texttt{Y}, q') \wedge \mathcal{R}_{\mathcal{L}}(i) \in \texttt{Own}(\texttt{X}, p') \wedge i_l \leq i \leq i_u \wedge q' \neq p'\} \\
\texttt{ProducersFor}(\texttt{S}, p) &= \{q' \mid \exists i \text{ such that } \mathcal{R}_{\mathcal{R}}(i) \in \texttt{Own}(\texttt{Y}, q') \wedge \mathcal{R}_{\mathcal{L}}(i) \in \texttt{Own}(\texttt{X}, p) \wedge i_l \leq i \leq i_u \wedge q' \neq p\} \\
\texttt{ConsumersFor}(\texttt{S}, q) &= \{p' \mid q \in \texttt{ProducersFor}(\texttt{S}, p')\} \\
\texttt{PutSet}(\texttt{S}, p, q) &= \{d \mid \exists i \text{ such that } d = \mathcal{R}_{\mathcal{R}}(i) \in \texttt{Own}(\texttt{Y}, q) \wedge \mathcal{R}_{\mathcal{L}}(i) \in \texttt{Own}(\texttt{X}, p) \wedge i_l \leq i \leq i_u \wedge q \neq p\} \\
\texttt{SendSet}(\texttt{S}) &= \{(q', p') \mid \exists d \text{ such that } d \in \texttt{PutSet}(\texttt{S}, p', q')\} \\
\texttt{Pending}(\texttt{S}) &= \{(p', q') \mid (q', p') \in \texttt{SendSet}(\texttt{S})\}
\end{aligned}
$$

The set Own(X,p) refers to the elements of array X mapped onto processor p through compiler directives. Similar Own sets are defined for other arrays as well. The sets Producers(S) and Consumers(S) denote, respectively, the processors that produce and consume data communicated in S. For a specific processor, ProducersFor and ConsumersFor give the set of processors that send data to and receive data from that processor respectively. PutSet(S,p,q) is the set of elements that should be put (written) by processor q to memory of processor p. SendSet(S) is set of pairs (q', p') such that q' sends data to (write data in the memory of) p'. Finally, Pending(S) is the inverse of SendSet(S), and gives set of pairs (p', q') such that p' should send a Synch message to q' for the repetitions of the communication (in each iteration of the time-step loop t) occurring in S. For a communication occurring in i, Pending(i) set represents a list of individual Synch messages that should be communicated for the safe repetition of the data communication in i. That is, a Synch message is between two processors only. For an i and a Synch, we can talk of whether or not

$$
\begin{aligned}
\texttt{Producers}(2) &= \{q \mid 1 \le q \le 7\} \\
\texttt{Consumers}(2) &= \{p \mid 0 \le p \le 6\} \\
\texttt{ProducersFor}(2,p) &= \{q \mid (q = p+1) \wedge (0 \le p \le 6)\} \\
\texttt{ConsumersFor}(2,q) &= \{p \mid (p = q-1) \wedge (1 \le q \le 7)\} \\
\texttt{PutSet}(2,p,q) &= \{d \mid (d = 16q) \wedge (p+1 = q) \wedge (1 \le q \le 7)\} \\
\texttt{SendSet}(2) &= \{(q, q-1) \mid 1 \le q \le 7\} \\
\texttt{Pending}(2) &= \{(p, p+1) \mid 0 \le p \le 6\}
\end{aligned}
$$

(a)

$$
\begin{aligned}
\texttt{Producers}(2) &= \{q \mid 0 \le q \le 7\} \\
\texttt{Consumers}(2) &= \{q \mid 0 \le q \le 7\} \\
\texttt{ProducersFor}(2,p) &= \{q \mid q = 0 \wedge p = 7\} \\
&\quad \cup \{q \mid (q = p+1) \wedge (0 \le p \le 6)\} \\
\texttt{ConsumersFor}(2,q) &= \{p \mid p = 7 \wedge q = 0\} \\
&\quad \cup \{p \mid (p = q-1) \wedge (1 \le q \le 7)\} \\
\texttt{PutSet}(2,p,q) &= \{d \mid \exists \alpha \text{ such that } (d = 32\alpha) \wedge (q = 0) \wedge (p = 7) \\
&\quad \wedge (32 \le d \le 128)\} \cup \\
&\quad \{d \mid \exists \alpha \text{ such that } (d = 4 + 4p + 32\alpha) \wedge (q = p+1) \\
&\quad \wedge (0 \le p \le 6) \wedge (4p+4 \le d \le 4p+100)\} \\
\texttt{SendSet}(2) &= \{(0,7)\} \cup \{(q, q-1) \mid 1 \le q \le 7\} \\
\texttt{Pending}(2) &= \{(p, p+1) \mid 0 \le p \le 6\} \cup \{(7,0)\}
\end{aligned}
$$

(b)

$$
\begin{aligned}
\{\vec{d} \mid \mathcal{P}(\vec{d})\} \vee_c \{\vec{d} \mid \mathcal{Q}(\vec{d})\} &= \{\vec{d} \mid \mathcal{P}(\vec{d}) \cup \mathcal{Q}(\vec{d})\} \\
\{\vec{d} \mid \mathcal{P}(\vec{d})\} -_c \{\vec{d} \mid \mathcal{Q}(\vec{d})\} &= \{\vec{d} \mid \mathcal{P}(\vec{d}) \cap \neg(\mathcal{Q}(\vec{d}))\} \\
\{\vec{d} \mid \mathcal{P}(\vec{d})\} \wedge_c \{\vec{d} \mid \mathcal{Q}(\vec{d})\} &= \{\vec{d} \mid \mathcal{P}(\vec{d}) \cap \mathcal{Q}(\vec{d})\}
\end{aligned}
$$

(c)

$$
\begin{aligned}
\texttt{PENDING_IN(i)} &= \bigsqcup_{j \in pred(i)} \texttt{PENDING_OUT(j)} \\
\texttt{PENDING_OUT(i)} &= \mathcal{F}_i(\texttt{PENDING_IN(i)})
\end{aligned}
$$

(d)

Figure 4: (a) Sets for the first loop in Figure 6(b) for `BLOCK` distribution. (b) Sets for the first loop in Figure 6(b) for `CYCLIC(4)` distribution. (c) Operations on `PutSets`. (d) Dataflow equations for optimizing synchronization messages.

$\texttt{Synch} \in \texttt{Pending(i)}$.

In fact, by using appropriate projection functions all of those sets can be obtained from a single set named `CommSet(S)` containing triples (p', q', d) meaning that element d should be communicated from q' to p' in `S`. The `CommSet(S)` is currently used in PARADIGM compiler's communication generation portion to generate `send` and `recv` commands. The necessary projection functions are implemented by using the Omega library [14, 11], and are shown in Figure 3(b). For instance, $\texttt{ConsumersFor(S}, q)$ is obtained from `CommSet(S)` by projecting out d and substituting q for q'; that is, q is a parameter and $\texttt{ConsumersFor(S}, q)$ enumerates p' values in terms of q.

By taking into account the alignment and distribution information, we can define `Own` set as

$$
\begin{aligned}
\texttt{Own(Y}, q) = \{d \mid \exists t, c, l \text{ such that } & (t = \alpha * d + \beta) \wedge (t = \mathcal{C} * P * c + \mathcal{C} * q + l) \wedge (y_l \le d \le y_u) \wedge \\
& (p_l \le q \le p_u) \wedge (t_l \le t \le t_u) \wedge (0 \le l \le \mathcal{C} - 1)\}
\end{aligned}
$$

where $P = p_u - p_l + 1$. In this formulation, $t = \alpha * d + \beta$ represents alignment information and $t = \mathcal{C} * P * c + \mathcal{C} * q + l$ denotes distribution information. In other words, each array element d is mapped onto a point in a two-dimensional array. This point can be represented by a pair (c, l) and gives the local address of the data item in a processor. Simple `BLOCK` and `CYCLIC(1)` distributions can be handled within this framework by setting $c = 0$ and $l = 0$, respectively. These relations can easily be modified to accommodate replicated arrays and collapsed dimensions. The formulation given here can be generalized to multi-dimensional loops, arrays and processor grids [2, 16]. As an example consider the first `i`-loop in Figure 6(b). Figure 4(a) shows the sets for this loop assuming that in the transformed program array bounds start from 0. Notice that a processor q is in the `Producers` set if there exists a processor p such that $q \neq p$ and q puts data in p's memory. Similarly, a processor p is in the `Consumers` set if there exists a processor q such that $p \neq q$ and q puts data in p's memory. For this example, if distribution directive for the arrays is changed to `CYCLIC(4)`, then we have the sets shown in Figure 4(b). All of these sets can easily be represented and manipulated by the Omega library [11, 14]. Notice that using the Omega sets to represent producer-consumer information, we are able to accommodate any kind of HPF-like alignment and distribution data in our framework through `Own` sets.

Let $\{\vec{d} \mid \mathcal{P}(\vec{d})\}$ and $\{\vec{d} \mid \mathcal{Q}(\vec{d})\}$ be two `PutSets` for a same multi-dimensional array where $\mathcal{P}(.)$ and $\mathcal{Q}(.)$ denote two predicates and $\vec{d}$ refers to an array element. We define three operations, $\vee_c$, $-_c$, and $\wedge_c$ on these `PutSets` as shown

in Figure 4(c). In the remainder of this paper, big $\vee$ and big $\wedge$ symbols will also be used for $\vee_c$ and $\wedge_c$ respectively.

4 Optimizing Communication

The objective of our global communication optimization framework is to determine $\texttt{PutSet}(\texttt{i}, p, q)$ for each node i in the program globally; that is, by taking into account all the nodes in the CFG that are involved in communication.

Local (Intra-Interval) Analysis In order to make the dataflow analysis task easier, the CFG of the program is traversed prior to the local analysis phase, and for each LHS reference a pointer is stored in the header of all enclosing loop nests. The local analysis part of our framework computes Kill, Gen and Post_Gen sets defined below for each interval. Then the interval is reduced to a single node and annotated with these sets. With reference to Figure 3(a),

$$\begin{aligned}
\texttt{Kill}(\texttt{i}, q) &= \{(\vec{d} \mid \vec{d} \in \texttt{Own}(\texttt{X}, q)) \wedge (\exists \vec{\imath} \text{ such that } (\vec{d} = \mathcal{R}_{\mathcal{L}}(\vec{\imath})) \wedge (\vec{i_l} \leq \vec{\imath} \leq \vec{i_u}))\}, \\
\texttt{Modified}(\texttt{i}, q) &= [\bigvee_{j \in pred(i)} \texttt{Modified}(\texttt{j}, q)] \vee_c \texttt{Kill}(\texttt{i}, q)
\end{aligned}$$

assuming that $\texttt{Modified}(\texttt{pred}(\texttt{first}(\texttt{i})), q) = \emptyset$ where first(i) is the first node in i.

$\texttt{Kill}(\texttt{i}, q)$ is the set of elements owned and written (killed) by processor q locally in i, and $\texttt{Modified}(\texttt{i}, q)$ is the set of elements that may be killed along any path from the beginning of the interval to (and including) i. The computation of the Kill set proceeds in the forward direction; that is, the nodes within the interval are traversed in topological sort order. If last(i) is the last node in i, then

$$\texttt{Kill}(\texttt{i}, q) = \texttt{Modified}(\texttt{last}(\texttt{i}), q)$$

This last equation is used to reduce an interval into a node. The reduced interval is then annotated by its Kill set.

$\texttt{Gen}(\texttt{i}, p, q)$ is the set of elements to be written by q into p's memory to satisfy the communication in i. The computation of the Gen proceeds in the backward direction, i.e., the nodes within each interval are traversed in reverse topological sort order. The elements that can be written by q into p's memory at the beginning of a node in the CFG are the elements required by p due to a RHS reference in the node except the ones that are written locally (killed) by q before being referenced by p.[1]

Assuming $\vec{\imath} = (\imath_1, ..., \imath_n)$ and $\vec{\imath'} = (\imath'_1, ..., \imath'_n)$, let $\vec{\imath'} \prec \vec{\imath}$ mean that $\vec{\imath'}$ is lexicographically less than or equal to $\vec{\imath}$; and $\vec{\imath'} \prec_k \vec{\imath}$ mean that $\imath'_j = \imath_j$ for all $j < k$, and $(\imath'_k, ..., \imath'_n) \prec (\imath_k, ..., \imath_n)$. Let $\texttt{Comm}(\texttt{i}, p, q)$ be the set of elements that may be communicated at the beginning of interval i to satisfy communication requirements from the beginning of i to the last node of i. Then, from Figure 3, assuming that $\texttt{Comm}(\texttt{succ}(\texttt{last}(\texttt{i})), q) = \emptyset$, we have

$$\begin{aligned}
\texttt{Gen}(\texttt{i}, p, q) &= \{\vec{d} \mid \exists \vec{\imath} \text{ such that } (\vec{i_l} \leq \vec{\imath} \leq \vec{i_u}) \wedge (\vec{d} = \mathcal{R}_{\mathcal{R}}(\vec{\imath}) \in \texttt{Own}(\texttt{Y}, q)) \wedge (\mathcal{R}_{\mathcal{L}}(\vec{\imath}) \in \texttt{Own}(\texttt{X}, p)) \\
&\quad \wedge \neg(\exists \vec{\jmath}, \mathcal{R}_{\mathcal{L}}{}' \text{ such that } (\vec{i_l} \leq \vec{\jmath} \leq \vec{i_u}) \wedge (\vec{d} = \mathcal{R}_{\mathcal{L}}{}'(\vec{\jmath})) \wedge (\vec{\jmath} \prec_{\texttt{level(i)}} \vec{\imath}))\}, \\
\texttt{Comm}(\texttt{i}, p, q) &= [\bigwedge_{s \in succ(i)} \texttt{Comm}(\texttt{s}, \texttt{p}, \texttt{q})] \vee_c \texttt{Gen}(\texttt{i}, \texttt{p}, \texttt{q}).
\end{aligned}$$

The negated condition eliminates all the elements written by q locally in an earlier iteration than the one in which p requires them. In addition, we use the following equation to reduce an interval into a single node

$$\texttt{Gen}(\texttt{i}, p, q) = \texttt{Comm}(\texttt{First}(\texttt{i}), p, q).$$

In the definition of Gen, $\mathcal{R}_{\mathcal{R}}$ denotes the RHS reference, and $\mathcal{R}_{\mathcal{L}}$ denotes the LHS reference of the same statement. $\mathcal{R}_{\mathcal{L}}{}'$, on the other hand, refers to any LHS reference within the same interval. Notice that while $\mathcal{R}_{\mathcal{L}}{}'$ is a reference to the same array as $\mathcal{R}_{\mathcal{R}}$, in general $\mathcal{R}_{\mathcal{L}}$ can be a reference to any array. level(i) gives the nesting level of the interval (loop), 1 corresponding to the outermost loop in the nest.[2]

After the interval is reduced, the Gen set for it is recorded, and an operator $\mathcal{N}$ is applied to the last part of this Gen

[1] Note that this process involves considering all LHS references within an interval. But since the largest scope for this analysis is a single loop nest and, as mentioned earlier, prior to analysis we keep pointers to all LHS references within a loop nest, this cost is manageable.

[2] In fact, there are more subtle issues here as lexical positions of the statements may also need be taken into account when Gen is computed. For the sake of simplicity, we omit the details.

Backward Analysis:

$$\texttt{Safe_Out}(\texttt{i},p,q) = \bigwedge_{s \in succ(i)} \texttt{Safe_In}(\texttt{s},p,q) \quad (1)$$

$$\texttt{Safe_In}(\texttt{i},p,q) = \begin{cases} \texttt{Gen}(\texttt{i},p,q) & \text{if } \mathcal{P}(\texttt{i}) \\ (\texttt{Safe_Out}(\texttt{i},p,q) -_c \texttt{Kill}(\texttt{i},q)) \vee_c \texttt{Gen}(\texttt{i},p,q) & \text{otherwise} \end{cases} \quad (2)$$

Forward Analysis:

$$\texttt{Put_In}(\texttt{i},p,q) = \bigwedge_{j \in pred(i)} \texttt{Put_Out}(\texttt{j},p,q) \quad (3)$$

$$\texttt{PutSet}(\texttt{i},p,q) = \begin{cases} \texttt{Gen}(\texttt{i},p,q) -_c \texttt{Put_In}(\texttt{i},p,q) & \text{if } \exists k \text{ such that } k \in succ(i) \text{ and } k \notin dom(i) \\ \texttt{Safe_In}(\texttt{i},p,q) -_c \texttt{Put_In}(\texttt{i},p,q) & \text{otherwise} \end{cases} \quad (4)$$

$$\texttt{Put_Out}(\texttt{i},p,q) = \begin{cases} \texttt{Put_In}(\texttt{i},p,q) -_c \texttt{Kill}(\texttt{i},q) & \text{if } \exists k \text{ such that } k \in succ(i) \text{ and } k \notin dom(i) \\ ((\texttt{PutSet}(\texttt{i},p,q) +_c \texttt{Put_In}(\texttt{i},p,q)) -_c \texttt{Kill}(\texttt{i},q)) +_c \texttt{Post_Gen}(\texttt{i},p,q) & \text{otherwise} \end{cases} \quad (5)$$

Figure 5: Dataflow equations for optimizing communication.

set to propagate it to the outer interval:

$$\mathcal{N}(\vec{j} \prec_{\texttt{k}} \vec{i}) = \vec{j} \prec_{(\texttt{k}-1)} \vec{i}.$$

It should be emphasized that computation of Gen sets gives us all the communication that can be vectorized above a loop nest; that is, our analysis easily handles message vectorization [9]. A naive implementation may set $\texttt{Put_Set}(\texttt{i},p,q)$ to $\texttt{Gen}(\texttt{i},p,q)$ for every i, p and q. But such an approach often retains redundant communication which would otherwise be eliminated.

Finally, $\texttt{Post_Gen}(\texttt{i},p,q)$ is the set of elements to be written by q into memory of p at node i with no subsequent local write to them by q:

$$\begin{aligned} \texttt{Post_Gen}(\texttt{i},p,q) = \; & \{\vec{d} \mid \exists \vec{\imath} \text{ such that } (\vec{i_l} \leq \vec{\imath} \leq \vec{i_u}) \wedge (\vec{d} = \mathcal{R}_{\mathcal{R}}(\vec{\imath}) \in \texttt{Own}(\texttt{X},q)) \wedge (\mathcal{R}_{\mathcal{L}}(\vec{\imath}) \in \texttt{Own}(\texttt{X},p)) \\ & \wedge \neg(\exists \vec{\jmath}, \mathcal{R}_{\mathcal{L}}{}' \text{ such that } (\vec{i_l} \leq \vec{\jmath} \leq \vec{i_u}) \wedge (\vec{d} = \mathcal{R}_{\mathcal{L}}{}'(\vec{\jmath})) \wedge (\vec{\imath} \prec_{\texttt{level(i)}} \vec{\jmath}))\}. \end{aligned}$$

The computation of $\texttt{Post_Gen}(\texttt{i},p,q)$ proceeds in the forward direction. Its computation is very similar to those of Kill and Gen sets, so we do not discuss it further.

Dataflow Equations In our framework, one way communication calls are placed at the beginning of nodes in the CFG. Our dataflow analysis consists of a backward and a forward pass. In the backward pass, the compiler determines sets of data elements that can safely be communicated at specific points. The forward pass, on the other hand, eliminates redundant communication and determines the final set of elements that should be communicated (written by q into p's memory) at the beginning of each node i. The input for the equations consists of the Gen, Kill and Post_Gen sets.

The dataflow equations for the backward analysis are given by Equations (1) and (2) in Figure 5. Basically, they are used to combine and hoist communication. $\texttt{Safe_In}(\texttt{i},p,q)$ and $\texttt{Safe_Out}(\texttt{i},p,q)$ are sets of elements that *can* safely be communicated at the beginning and end of node i, respectively. Equation (1) says that an element should be communicated at a point if and only if it will be used in all of the following paths in the CFG. Equation (2), on the other hand, gives the set of elements that can safely be communicated at the beginning of i. Intuitively, an element can be written by q into p's memory at the beginning of i if and only if it is either required by p in i or it reaches at the end of i and is not overwritten (killed) by the owner (q) in it. The predicate $\mathcal{P}(\texttt{i})$ is used to control communication hoisting. If $\mathcal{P}(\texttt{i})$ is true, communication is not hoisted to the beginning of i. $\mathcal{P}(\texttt{i})$=*false* implies aggressive communication combining and hoisting. An algorithm can also put a condition which tests the compatibility between $\texttt{Gen}(\texttt{i},p,q)$ and $\texttt{Safe_Out}(\texttt{i},p,q)$ (e.g. two left-shift communications are compatible whereas a left-shift and a right-shift are not) [5]. Among the other possible alternatives are avoiding message splitting ($\texttt{Kill}(\texttt{i},q) \neq \emptyset$), clustering ($\texttt{Gen}(\texttt{i},p,q) = \emptyset$) and avoiding the buffer pressure [12]. In the experiments, except tomcatv, we combined and hoisted communication

aggressively. For tomcatv, we checked for compatibility before combining. Note that the redefinition of $\mathcal{P}$(i) can totally change the behavior of the algorithm as well as the empirical results.

The task of the forward analysis phase, which makes use of Equations (3), (4) and (5) in Figure 5, is to eliminate redundant communication by observing that (1) a node in the CFG should not have a non-local datum which is exclusively needed by a successor unless it dominates that successor; and (2) a successor should ignore what a predecessor has so far unless that predecessor dominates it. Put_In(i, p, q) and Put_Out(i, p, q) denote the set of elements that *have been* written so far (at the beginning and end of node i respectively) by q into memory of p. Equation (3) conservatively says that the communication set arriving in a join node can be found by intersecting the sets for all the joining paths. Equation (4) is used to compute the PutSet set which corresponds to the elements that can be communicated at the beginning of the node except the ones that have already been communicated (Put_In). The elements that have been communicated at the end of node i (that is, Put_Out set) are simply the union of the elements communicated up to the beginning of i (that is, Put_In set), the elements communicated at the beginning of i (that is, PutSet(i,p,q) set) (except the ones which have been killed in i) and the elements communicated in i and not killed subsequently (Post_Gen).

Interval Analysis Our approach starts by computing the Gen, Kill and Post_Gen sets for each node. Then the contraction phase of the analysis reduces the intervals from the innermost loop to the outermost loop and annotates them with Gen, Kill and Post_Gen sets. When a reduced CFG with no cycles is reached, the expansion phase starts and PutSets for each node is computed from the outermost loop to the innermost loop. There is one important point to note: before starting to process the next graph in the expansion phase, the Put_In set of the first node in this graph is set to the PutSet of the interval that contains it to avoid redundant communication. More formally, in the expansion phase, we set $\texttt{Put_In}(\texttt{i}, p, q)^{k^{th}pass} = \texttt{PutSet}(\texttt{i}, p, q)^{(k-1)^{th}pass.}$. This assignment then triggers the next pass in the expansion phase. Before the expansion phase starts $\texttt{Put_In}(\texttt{i}, \texttt{p}, \texttt{q})^{1^{st}pass}$ is set to the empty set. Note that the whole dataflow procedure operates on sets of equalities and inequalities which can be manipulated by the Omega library [11] or a similar tool. A more detailed discussion of the Gen, Kill and Post_Gen sets as well as example computations can be found in [10].

Example Consider the synthetic benchmark given in Figure 6(a). In this example, communication occurs for three arrays: B, D and F. A communication optimization scheme based on message vectorization alone, can place communications and associated synchronizations as shown in Figure 6(b) before the loop bounds reduction. Note that a Synch message in that figure in fact represents a number of point-to-point synchronization messages. In particular, for a communication occurring at i, a processor p must get synchronized with every q that satisfies $(p, q) \in$ Pending(i). An application of our global communication optimization method generates the program shown in Figure 6(c). As compared with the message vectorized version, there is a 50% reduction (from 28 to 14) in the number of messages and 40% in the communication volume (from 35 to 21) across all processors. We note that we can optimize this program even the distribution directive is changed to CYCLIC(K) for any K. None of the previous approaches is able to handle the global communication optimization for the case K $\geq$ 2, mostly due to the representations they use for communication sets. In fact, when the distribution directive is CYCLIC(4), we have a 50% reduction (from 48 to 24) in the number of messages and 32% reduction (from 139 to 94) in the communication volume across all processors. Note that our approach here reduces the original number of synchronization messages as well (from 28 to 14 for the BLOCK distribution and from 48 to 24 for the CYCLIC(4) case). We investigate the conditions under which the synchronization messages can be further decreased in the following section.

5 Optimizing Synchronization

In this section, we assume that the compiler has conducted the dataflow analysis described in Section 4 and determined the optimal communication points and communication sets. Assuming that these communications will be implemented by Put operations, we present a dataflow analysis to minimize the number of Synch messages. We assume that communication patterns (i.e. producer-consumer relationships) are *identical* for each repetition of a communication during the loop execution. For example, in Figure 6(c), the producer-consumer pattern for the communication occurring in 1 is identical for every repetition of time-step loop t. The overall algorithm for eliminating redundant Synch messages is given in Figure 7. The routines called by this algorithm are explained in this section.

```
 REAL A(128), B(128), C(128)
     D(128), E(128), F(128)
!HPF$ PROCESSORS PROC(0:7)
!HPF$ DISTRIBUTE BLOCK ON TO
     PROC ::  A, B, C, D, E

  DO t = 1, TIMES
   DO i = 1, 127
S:  A(i) = B(i+1)
   END DO
   DO i = 1, 126
    D(i) = A(i) + 1
    A(i) = B(i+1) + B(i+2)
   END DO
   IF (A(1).EQ.B(1))
    DO i = 2, 64
     C(i) = D(i-1)
    END DO
   ELSE
    DO i = 2, 64
     A(i) = C(i) + D(i-1)
    END DO
   END IF
   DO i = 1, 127
    F(i) = A(i) - 1
    B(i) = A(i) * A(i)
   END DO
   DO i = 65, 128
    E(i) = F(i-1)
   END DO
  END DO
       (a)
```

```
 REAL A(128), B(128), C(128)
     D(128), E(128), F(128)
!HPF$ PROCESSORS PROC(0:7)
!HPF$ DISTRIBUTE BLOCK ON TO
     PROC ::  A, B, C, D, E

1 DO t = 1, TIMES
2  Synch; put{B};
3  DO i = 1, 127
4   A(i) = B(i+1)
5  END DO
6  Synch; put{B};
7  Synch; put{B};
8  DO i = 1, 126
9   D(i) = A(i) + 1
10  A(i) = B(i+1) + B(i+2)
11 END DO
12 IF (A(1).EQ.B(1))
13  Synch; put{D};
14  DO i = 2, 64
15   C(i) = D(i-1)
16  END DO
17 ELSE
18  Synch; put{D};
19  DO i = 2, 64
20   A(i) = C(i) + D(i-1)
21  END DO
22 END IF
23 DO i = 1, 127
24  F(i) = A(i) - 1
25  B(i) = A(i) * A(i)
26 END DO
27 Synch; put{F};
28 DO i = 65, 128
29  E(i) = F(i-1)
30 END DO
31END DO
        (b)
```

```
 REAL A(128), B(128), C(128)
     D(128), E(128), F(128)
!HPF$ PROCESSORS PROC(0:7)
!HPF$ DISTRIBUTE BLOCK ON TO
     PROC ::  A, B, C, D, E

  DO t = 1, TIMES
1: Synch; put{B};
   DO i = 1, 127
    A(i) = B(i+1)
   END DO
   DO i = 1, 126
    D(i) = A(i) + 1
    A(i) = B(i+1) + B(i+2)
   END DO
2: Synch; put{D};
   IF (A(1).EQ.B(1))
    DO i = 2, 64
     C(i) = D(i-1)
    END DO
   ELSE
    DO i = 2, 64
     A(i) = C(i) + D(i-1)
    END DO
   END IF
   DO i = 1, 127
    F(i) = A(i) - 1
    B(i) = A(i) * A(i)
   END DO
3: Synch; put{F};
   DO i = 65, 128
    E(i) = F(i-1)
   END DO
  END DO
       (c)
```

Figure 6: (a) a synthetic benchmark. (b) message vectorized translation. (c) global communication optimization.

Our approach first makes a single pass over the current interval and determines some synchronizations that cannot be eliminated by the analysis to be described. Informally, if a synchronization is repeated in the program in different places with the same producer-consumer and is associated with (the part of a same) buffer and there is no intervening communication in the synchronization direction, then that synchronization cannot be eliminated. We call the set of synchronizations (associated with a node `i`) that cannot be eliminated `SynchFix(i)`. The algorithm `SYNC_FIX` given in Figure 8 performs this pre-processing function.

The dataflow technique described here starts from the deepest loops and works its way through loops in a bottom up manner, handling one loop at a time. It then reduces the loop to a node and annotates it with its final synchronization requirements which cannot be eliminated. Since the approach starts from the deepest loops, it tends to eliminate the synchronization requirements from the most frequently executed parts of the program first. The synchronization is introduced at the least executed parts of the code only to preserve the correct semantics. The procedure works on an augmented CFG, where each communication loop is represented by a single node. In the following, the symbol `i` refers to such a node.

For a given `i`, we can define the set `SynchSet` as a set of processor pairs that should be synchronized after our analysis. In a straightforward implementation, `SynchSet(i)` = `Pending(i)` for each `i`. We would like to reduce the cardinality of `SynchSet(i)` for each `i`. We say that no synchronization is required for a communication `i`, if `SynchSet(i)` happens to be empty after our dataflow analysis.

- INPUT: *The CFG after global communication optimization.*
- OUTPUT: *The CFG where redundant* `Synch` *messages are eliminated.*

```
foreach Interval ∈ CFG
  Pending(Interval) ← ∅
  SendSet(Interval) ← ∅
  for i = 1, lastnode(Interval)
    compute Pending(i)
    Pending(Interval) ← Pending(Interval) ∪ Pending(i)
    compute SendSet(i)
    SendSet(Interval) ← SendSet(Interval) ∪ SendSet(i)
  endfor
  SynchFix(Interval) ← SYNC_FIX(Pending(Interval), PutSet(k) | ∀k)
  Pending(Interval) ← SYNC_ELIM(Pending(Interval),SendSet(Interval))
  Pending(Interval) ← SYNC_FOR_SYNC(Pending(Interval))
  SynchSet(Interval) ← Pending(Interval) ∪ SynchFix(Interval)
  reduce SynchSet(Interval) into a single set
endforeach
```

Figure 7: The synchronization elimination algorithm.

If the compiler wants to eliminate a `Synch` message for communication `i` from p to q, it needs to find a message $\mathtt{t}_j$ for another communication from p to q and use it as synchronization. Such a message should occur *between* repetitions of `i` and *after* the data value communicated at `i` is consumed. Suppose that a specific producer q and a consumer p are involved in a communication in `i`. Consider all k communications $\mathtt{t}_j$ $(1 \leq j \leq k)$ occurring after the value communicated in `i` is consumed by p and before the next repetition of `i`. Then, if the following holds, the `Synch` message from p to q can be eliminated:

$$(\exists j \mid (1 \leq j \leq k) \wedge q \in \mathtt{ConsumersFor}(\mathtt{t}_j, p))$$

An interesting case occurs when all the `Synch` messages contained in the `Pending(i)` set for a specific `i` are eliminated. We can formalize this condition as

$$\forall p \forall q(((p,q) \in \mathtt{Pending(i)}) \quad \Rightarrow \quad \exists j((p,q) \in \mathtt{SendSet}(\mathtt{t}_j))) \tag{6}$$

assuming $1 \leq j \leq k$. Notice that j values can be different for each q. If we additionally stipulate that all j values should be the same for all q values, then we obtain

$$\forall p \forall q \exists j(((p,q) \in \mathtt{Pending(i)}) \quad \Rightarrow \quad ((p,q) \in \mathtt{SendSet}(\mathtt{t}_j))) \tag{7}$$

We note that the condition (7) implies the algorithms offered by [6] and [8].

Claim: *The synchronizations eliminated by (7) are a subset of the synchronizations that can be eliminated by (6).*

Proof: Let the set of synchronizations eliminated by scheme (6) be S_1 and those eliminated by scheme (7) be S_2. We need to show that $S_2 \subseteq S_1$.

Consider a synchronization, $(p,q) \in S_2$. Since $(p,q) \in S_2$, condition (7) above implies that $\exists j$ s.t. $q \in ConsumersFor(t_j, p)$. From condition (6) above this implies that $(p,q) \in S_1$.

Now consider a case where there are two synchronization requirements (p_1,q_1) and (p_2,q_2). Let $j_1, j_2 (j_1 \neq j_2)$ be such that $q_1 \in ConsumersFor(t_{j_1}, p_1)$ and $q_2 \in ConsumersFor(t_{j_2}, p_2)$ and there is no other j which meets this condition. By condition (6),
$\{(p_1,q_1),(p_2,q_2)\} \in S_1$. However, since $j_1 \neq j_2$, $\{(p_1,q_1),(p_2,q_2)\} \notin S_2$ by condition (7). Hence $S_2 \subseteq S_1$.

- INPUT: `Pending(i)` *and* `PutSet(i)` | ∀`i`.
- OUTPUT: *A set of synchronizations that should not be eliminated.*

for `node` = 1, lastnode(*Interval*)
  foreach $(p, q) \in$ `Pending(node)`
    if (∃ `j` such that
      $(p, q) \in$ `Pending(j)` and `j` $<$ `node` and
      `PutSet(node)` $\cap$ `PutSet(j)` $\neq \emptyset$ and
      $\neg$ (∃ `k` such that `j` $<$ `k` $<$ `node` and
      $(p, q) \in$ `SendSet(k)`)) then
        `SynchFix(node)` $\leftarrow$ `SynchFix(node)` $\cup\{(p, q)\}$
    endif
  endforeach
endfor
return `SynchFix(i)` | ∀ i

Figure 8: `SYNC_FIX` routine.

Even if a `Pending(i)` set cannot be totally eliminated we can reduce its cardinality by eliminating as many `Synch` messages as possible from it. That is, after our analysis, for every `i`,

$$\texttt{SynchSet(i)} \quad = \quad \texttt{Pending(i)} -_c \{(p, q) \mid \exists j \text{ such that } q \in \texttt{ConsumersFor}(\texttt{t}_j, p)\} \vee_c \texttt{SynchFix(i)} \tag{8}$$

As an example, let us consider the program shown in Figure 6(c). In this program, communication occurs at three points: `1`, `2` and `3`. A straightforward implementation inserts three (sets of) `Synch` operations, corresponding to `1`, `2` and `3` as shown in that figure. Let us now focus on the communication in `1`. In fact, this communication is combination of communications due to references `B(i+1)`, `B(i+1)` and `B(i+2)` in lines 4 and 10 in Figure 6(b). Figure 2(a) shows the messages sent (`Put` operations) for this communication. Figure 2(b), on the other hand, shows the required synchronization messages for the repetitions of this communication. Finally, Figures 2(c) and 2(d) show the communication messages in `2` and `3` respectively. Notice that, by using the condition given in (6), the synchronization requirements for `1` can be eliminated; that is, the communications occurring in `2` and `3` together kill the synchronization requirements for the repetitions of the communication in `1`. If we consider Figure 2(c) and Figure 2(d) separately for the condition given by (7), however, none of them individually can eliminate the synchronization for `1`. By a similar argument, it can be concluded that the communications in `2` and `3` do not need any synchronization either, as their synchronization requirements are killed by communication in `1`.

Dataflow Analysis: We now present our dataflow analysis for eliminating redundant synchronization messages. The data flow equations are shown in Figure 4(d). Our analysis consists of iterative forward passes on the augmented CFG. Let us first concentrate on the second equation in that figure and explain the functionality of $\mathcal{F}_i$. In that equation, `PENDING_IN(i)` represents the synchronization requirements of all the communications traversed so far up to the beginning of `i`. `PENDING_OUT(i)` is defined analogously for the end of `i`.

Assuming that P_k is the synchronization requirement (in terms of pairs of processors) for node `k` up to `i` in the analysis and `PENDING_IN(i)` = $\{P_1, P_2, ..., P_{i-1}, P_i, P_{i+1}, ..., P_m\}$, we can define $\mathcal{F}_i$ as

$$\mathcal{F}_i(\texttt{PENDING_IN(i)}) \quad = \quad \{f_i(P_1), f_i(P_2), ..., f_i(P_{i-1}), P_i, f_i(P_{i+1}), ..., f_i(P_m)\}$$

where $f_i(P_k) = P_k -_c$ `SendSet(i)` . That is, when a node `i` is visited, the synchronization requirements for all other communications are checked to see whether or not any synchronization can be eliminated by using the communication occurring at `i`. Prior to the analysis, P_i is set to `Pending(i)` for the *first* node. After a fixed state is reached, `PENDING_OUT` set of the *last* node gives the synchronization requirements to be satisfied. In fact, except for the statements in the conditionally executed blocks, in steady state, all `PENDING_OUT` sets should be the same. The resulting

PENDING_OUT set, which is in fact a set of sets, is then reduced to a single set and used to represent the synchronization requirements of this loop to the next upper level.

We now consider the first equation of Figure 4(d) and explain the big $\sqcup$ operator appearing there. In the join nodes, the compiler takes a conservative approach by unioning the synchronization requirements for the same communication. Suppose that for each $j \in pred(i)$, PENDING_OUT(j) = $\{P_{j1}, P_{j2}, ..., P_{jm}\}$, where P_{jk} is the synchronization requirement of communication k up to the end of j. Assuming then that the resulting PENDING_IN(i) = $\{P_1, P_2, ..., P_m\}$, each $P_l \in$ PENDING_IN(i) can be computed as $P_l = \vee_c P_{jl}$ where $\vee_c$ is performed over the j values. We note that this algorithm is more accurate and faster than those proposed in [8] and [6].

Claim: *The dataflow procedure defined by equations given in Figure 4(d) can reach a steady state in at most after two iterations.*

> **Proof:** Consider an arbitrary statement i in the interval which needs synchronization (p, q). This can be eliminated by a communication (p, q) in any statement t_j by condition (1). There are two possibilities.
>
> *Case 1. $j < i$:* Since we are doing a forward pass over the interval, we would visit node j sometime in the first iteration itself and eliminate (p, q)
>
> *Case 2. $j > i$:* We will miss node j in the first iteration, but surely we will visit it in the second iteration.

Example The upper part of Table 1 shows the SendSet and Pending sets (in an open form rather than in terms of equalities and inequalities) for the program shown in Figure 6(c). Note that SendSet(1) and Pending(1) are computed from the SendSet and Pending sets respectively given in Figure 4(a). The sets for 2 and 3 are computed similarly. Before the dataflow analysis starts, PENDING_IN(1) is initialized as follows:

$$\begin{aligned} \texttt{PENDING_IN(1)} &= \{\texttt{Pending(1)}, \texttt{Pending(2)}, \texttt{Pending(3)}\} \\ &= \{\{(0,1),(1,2),(2,3),(3,4),(4,5),(5,6),(6,7)\},\{(1,0),(2,1),(3,2)\},\{(4,3),(5,4),(6,5),(7,6)\}\} \end{aligned}$$

The lower part of Table 1 demonstrates application of our dataflow algorithm to this example. After the fixed state is reached, an examination of PENDING_OUT(3) reveals that the program can be executed without any synchronization.

Our synchronization elimination algorithm SYNC_ELIM and the algorithm which computes the $\mathcal{F}$ operator are given in Figures 9 and 10 respectively.

Eliminating Synch messages by other Synch messages In some cases, when all communication is in the same direction, Synch messages cannot be eliminated by data messages. However, they can still be eliminated by "other" Synch messages. In our experiments, this happened with stfrg benchmark whose message vectorized version contains six similar communications. The algorithm that we use for eliminating Synch messages with Synch messages is rather straightforward. We check all possible subsets of Synch messages and determine how many Synch messages are eliminated if we opt to use that subset. We then choose the subset which eliminates the most number of messages. In case of tie, we choose the alternative with the least number of elements, in an attempt to reduce the code complexity. Note that although this approach requires an exhaustive search (i.e., it is exponential in number of Pending sets), since a portion of the Synch messages is already eliminated by global communication analysis, the cost is manageable. Our algorithm (SYNC_FOR_SYNC) for eliminating Synch messages using other Synch messages is shown in Figure 11.

Varying producer-consumer patterns In the following, we concentrate on eliminating synchronization messages when producer-consumer relationship varies with loop index variable. In general it is difficult to eliminate the synchronization messages in this case. Here we formulate the problem, and present the conditions under which the synchronization messages can be eliminated.

Consider Figure 12 assuming that (a), (b) and (c) correspond to communication patterns of three iterations of a loop. After (a), in order to eliminate the Synch message required for (b), data message shown in (d) should be taken. Similarly, after (b), data message shown in (e), and after (c), data message shown in (f) should be taken. In general, each data message t_i should kill the synchronization requirement of the communication that will occur in the next iteration of the loop.

Let us now define set $\mathcal{I}' = \{i \mid (\mathcal{R}_{\mathcal{L}}(i)$ is owned by $p) \wedge (\mathcal{R}_{\mathcal{R}}(i)$ is owned by $q) \wedge (p \neq q)\}$; i.e. the set of iterations that require communication. For the sake of explanation, we assume a two-deep loop nest where the producer-consumer relationship is variant within $\imath$–loop which is outermost; but is constant for $\jmath$–loop which is innermost, assuming $\tau = (\imath, \jmath)^T$. We define

- INPUT: *Two sets comprising of* `Pending(i)` *and* `SendSet(i)` *sets respectively for each* `i`.
- OUTPUT: *A set representing the remaining synchronization requirements after the synchronization elimination.*

```
PENDING_IN(1) ← ∅
for i = 1, lastnode(Interval)
  PENDING_IN(1) ← PENDING_IN(1) ∪ Pending(i)
endfor
change ← true
while change
  change ← false
  OLDPENDING ← PENDING_IN(1)
  PENDING_OUT(1) ← F(1,PENDING_IN(1))
  for i = 2, lastnode(Interval)
    PENDING_IN(i) ← ⊔_{j∈pred(i)} PENDING_OUT(j)
    PENDING_OUT(i) ← F(i,PENDING_IN(i))
  endfor
  PENDING_IN(1) ← PENDING_OUT(lastnode(Interval))
  if PENDING_IN(1) ≠ OLDPENDING then
    change ← true
  endif
endwhile
return PENDING_OUT(lastnode(Interval))
```

Figure 9: `SYNC_ELIM` routine.

$$\begin{aligned}
\texttt{SendSet}(\texttt{i},\imath) &= \{(p,q) \mid \exists \jmath \text{ such that } (\tau \in \mathcal{I}') \wedge (\mathcal{R}_{\mathcal{L}}(\tau) \in \texttt{Own}(\texttt{X},p)) \wedge (\mathcal{R}_{\mathcal{R}}(\tau) \in \texttt{Own}(\texttt{Y},q)) \wedge (\jmath_l \leq \jmath \leq \jmath_u) \\
&\quad \wedge (p \neq q) \wedge (p_l \leq p, q \leq p_u)\} \\
\texttt{Pending}(\texttt{i},\imath) &= \{(p,q) \mid (q,p) \in \texttt{SendSet}(\texttt{i},\imath)\}
\end{aligned}$$

We now define two order relations, ∇ and ∇_s, on the universe of `Pending` sets.

$$\texttt{Pending}(\texttt{i},\imath) \nabla \texttt{Pending}(\texttt{i},\imath') \text{ if and only if } \imath_l \leq \imath < \imath' \leq \imath_u$$

and

$$\texttt{Pending}(\texttt{i},\imath) \nabla_s \texttt{Pending}(\texttt{i},\imath') \text{ if and only if}$$

$$(\texttt{Pending}(\texttt{i},\imath) \nabla \texttt{Pending}(\texttt{i},\imath')) \wedge \text{ not } (\exists \imath'' \text{ such that } (\texttt{Pending}(\texttt{i},\imath) \nabla \texttt{Pending}(\texttt{i},\imath'')) \wedge (\texttt{Pending}(\texttt{i},\imath'') \nabla \texttt{Pending}(\texttt{i},\imath')))$$

Note that ∇_s orders the $\texttt{Pending}(\texttt{i},\imath)$ sets according to increasing values of loop index $\imath$. We say that a $\texttt{Synch}(\texttt{i})$ set can totally be eliminated if the following holds.

$$\begin{aligned}
\forall p \forall q \forall \imath (((p,q) \in \texttt{Pending}(\texttt{i},\imath)) \wedge (\texttt{Pending}(\texttt{i},\imath) \nabla_s \texttt{Pending}(\texttt{i},\imath')) \wedge ((p',q') \in \texttt{Pending}(\texttt{i},\imath'))) \\
\Rightarrow \exists j ((p',q') \in \texttt{SendSet}(\texttt{t}_j,\imath))) \qquad (9)
\end{aligned}$$

Notice that the formulation given here can be generalized for the case in which there are n_1 loops before the commu-

- INPUT: node*: a node number;* PENDING=$\{P_1, P_2, ...\}$ *where* PENDING(k) $= P_k$ *is the synchronization requirement for node k.*
- OUTPUT: *An updated* PENDING *set.*

```
for k = 1, lastnode(Interval)
  if k ≠ node then
    PENDING(k) ← PENDING(k) − SendSet(node)
  endif
endfor
return PENDING
```

Figure 10: $\mathcal{F}$ routine.

Table 1: Dataflow sets and PENDING_OUT sets for the example in Figure 6(c).

communication in	SendSet	Pending
1	{(1,0),(2,1),(3,2),(4,3),(5,4),(6,5),(7,6)}	{(0,1),(1,2),(2,3),(3,4),(4,5),(5,6),(6,7)}
2	{(0,1),(1,2),(2,3)}	{(1,0),(2,1),(3,2)}
3	{(3,4),(4,5),(5,6),(6,7)}	{(4,3),(5,4),(6,5),(7,6)}

PENDING_OUT for	iteration 1	iteration 2
1	$\{\{(0,1),(1,2),(2,3),(3,4),(4,5),(5,6),(6,7)\},\{\emptyset\},\{\emptyset\}\}$	$\{\{\emptyset\},\{\emptyset\},\{\emptyset\}\}$
2	$\{\{(3,4),(4,5),(5,6),(6,7)\},\{\emptyset\},\{\emptyset\}\}$	$\{\{\emptyset\},\{\emptyset\},\{\emptyset\}\}$
3	$\{\{\emptyset\},\{\emptyset\},\{\emptyset\}\}$	$\{\{\emptyset\},\{\emptyset\},\{\emptyset\}\}$

nication and n_2 loops after the communication; consequently it is more powerful than the one offered by Gupta and Schonberg in [6].

6 Preliminary Results

The applications used in our study and their characteristics are listed in Table 2. We experimented with BLOCK, B-CYC (block-cyclic) and CYCLIC distributions on 8 and 32 processors to measure the static improvements in communication volume, number of data messages and number of synchronization messages. For SOR, we used a parallelizable version which uses two arrays. In swim256, no in-lining has been performed. We refer to a version of program which is optimized by message vectorization alone as base. Table 3 presents the communication volume and the number of data messages (that is also the number of Synch messages) in the base programs.

For these applications, we first applied our global communication optimization algorithm. The percentage improvements are listed in Table 4. We note that for tomcatv, ADI and SOR, the global optimizer is not able to improve over the base version.[3] For vpenta, the optimizations were able to reduce the number of messages only. In particular, in block-cyclic distributions where the previous approaches fail, we have, on the average, a 29% reduction in communication volume and 40% reduction in number of messages across all processors.

We then applied our synchronization elimination scheme to the base version as well as the globally optimized version (C-opt) for each program. The results shown in Table 5 reveal that the algorithm is surprisingly successful in eliminating the redundant Synch messages. Except for two programs (stfrg and swim256), the algorithm eliminates

[3] That means the algorithm is unable to reduce the number of messages or the communication volume. However, our code generator was able to reduce the loop overhead somewhat.

- INPUT: `Pending(i)` | $\forall$`i` = *set of the synchronizations that cannot be eliminated by* `SYNC_ELIM`.
- OUTPUT: *A selected set of synchronizations that should be used to eliminate the remaining synchronizations.*

```
PENDINGSOFAR ← ∅
for i = 1, lastnode(Interval)
  PENDINGSOFAR ← PENDINGSOFAR ∪ Pending(i)
endfor
SELECTED ← ∅
min_size ← 0
max_eliminated ← 0
foreach subset K of PENDINGSOFAR
  temp ← K
  temp ← SYNC_ELIM(PENDINGSOFAR-temp,temp)
  determine eliminated
  if eliminated > max_eliminated then
    SELECTED ← K
    max_eliminated ← eliminated
    min_size ← |K|
  else if eliminated = max_eliminated then
    if |K| < min_size then
      min_size ← |K|
      SELECTED ← K
    endif
  endif
endforeach
return SELECTED
```

Figure 11: `SYNC_FOR_SYNC` routine.

all synchronization messages from the `base` programs. In `stfrg`, since all the communication in the same direction, it uses synchronization messages to eliminate other synchronization messages. In `swim`, the synchronizations that could not be eliminated all occur in the initialization subroutine. When we look at the results for the programs that are optimized for communication prior to synchronization analysis, however, the picture somewhat changes. In `C-Opt` versions of `Jacobi` and `stfrg`, since the communication loop is reduced to one, no synchronization messages can be eliminated. For the other programs, a large portion of the synchronization messages is eliminated.

To sum up, our communication optimization algorithm eliminates 37.3% of the data messages and synchronization messages and reduces communication volume across all processors by 26%. Our synchronization elimination algorithm eliminates 96.8% of the synchronization messages in the message vectorized programs and 74% of the synchronization messages in the globally optimized programs.

7 Related Work

Previously several methods have been presented to optimize the communication on distributed-memory machines. Most of the efforts considered communication optimization at loop level. Although each approach has its own unique features, the general idea is to apply an appropriate combination of message vectorization, message coalescing and message aggregation [20, 9, 3].

More recently some researchers have proposed techniques based on dataflow analysis in order to optimize communication across multiple loop nests. The works in [7], [12], [5] and [19] present similar frameworks to optimize `send`/`recv` communication globally. Almost all of the approaches use a variant of Regular Section Descriptors (RSD)

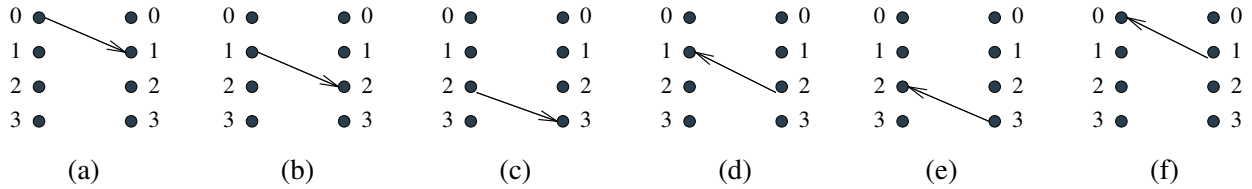


Figure 12: An example communication pattern.

introduced by Callahan and Kennedy [4]. For each array referenced in the program, an RSD is defined which describes the portion of the array that is referenced. Although this representation is convenient for simple array sections such as those found in pure block or cyclic distributions, it is hard to embed alignment and general distribution information into it. Apart from failing to support block-cyclic distributions, working with section descriptors may sometimes result in overestimation of the communication sets. The inaccuracy resulting from these estimations may be linear with the number of dataflow formulations to be evaluated, thus defeating the purpose of global communication optimization. Instead, our approach is based on a linear algebra framework, and can represent all HPF-like alignment and distribution styles uniformly. It should be emphasized that our communication optimization approach can also be used for optimizing `send/recv` type of communication on distributed-memory machines.

The approaches given in [6] and [8] examine the problem of eliminating redundant synchronization operations by piggy-backing them on data messages. While [8] uses barrier synchronizations, [6] focuses on point-to-point synchronization messages. Our approach is superior to both of these in eliminating synchronization as explained in the paper.

Subhlok [17] addresses the problem of synchronization in Fortran programs with parallel loop constructs. His approach determines whether the current synchronization points in the program are sufficient. Instead, we employ communication analysis which even in shared memory systems can aggressively optimize synchronization.

O'Boyle and Bodin [13] and Tseng [18] focus on synchronization elimination problem on shared virtual memory and distributed shared memory systems respectively. There is an important difference between our work and theirs. They eliminate synchronizations which are introduced by the insufficient communication analysis performed by the shared memory compilers. A compiler approach based on distributed-memory paradigm (like ours) does not insert those synchronizations at the first place. In our case, we start with an unoptimized program in which those types of artificial synchronizations are non-existent anyway. We rather focus on elimination of synchronizations that are caused by mandatory data communications. Those types of synchronizations are not eliminated by neither [13] nor [18]; though [18] replaces barriers with software-implemented counters in specific cases.

8 Summary

We presented dataflow algorithms to reduce number of data messages, communication volume and number of synchronization messages. Our experimental results revealed that our approach is quite successful in practice reducing on average 37.3% of data messages and 96.8% of synchronization messages in the message vectorized programs.

Our future work along this direction will include compiling data-parallel programs with `Get` primitives, and elimination of synchronizations and possible deadlocks from programs compiled under hybrid approaches which employ both `Put/Get` and `send/recv` primitives.

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Table 2: Programs in our experiment set and their characteristics (+ = block, block-cyclic(4) or cyclic; * = not distributed).

PROGRAM	SUITE	LINES	ARRAYS	SIZE	DISTRIBUTION	DESCRIPTION
Jacobi	–	25	$2 \times 2D$	1024×1024	(+,*)	Jacobi iteration
2D hydro	Livermore (18)	38	$9 \times 2D$	512×512	(+,*)	2D hydrodynamics
ADI	Livermore (8)	30	$3 \times 3D,\ 3 \times 1D$	$256 \times 256 \times 2$	(+,*,*)	ADI fragment
vpenta	Spec92/NAS	147	$2 \times 3D,\ 7 \times 2D$	$128 \times 128 \times 3$	(*,+,*)	pentadiagonal inversion
SOR	–	25	$2 \times 2D$	256×256	(+,*)	successive over-relaxation
stfrg	Livermore (7)	17	$1 \times 1D$	1024	(+)	state fragment equation
tomcatv	Spec95	190	$7 \times 2D,\ 2 \times 1D$	512×512	(+,*)	2D mesh generation
swim256	Spec95	428	$14 \times 2D$	512×512	(+,*)	shallow water equation solver

Table 3: Communication volume and number of data messages in the `base` (message vectorized) programs for a single iteration of the time-step loop.

PROGRAM	communication volume (in thousand elements)						number of data messages (Synch messages)					
	Number of Processors = 8			Number of Processors = 32			Number of Processors = 8			Number of Processors = 32		
	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC
Jacobi	14.4	522.2	2,088.9	63.3	522.2	2,088.9	14	16	16	62	64	64
2D hydro	57.1	1,036.3	4,169.8	252.9	1,036.3	4,169.8	112	128	128	496	512	512
ADI	10.7	96.8	387.1	47.2	96.8	387.1	42	48	48	186	192	192
vpenta	343.2	343.2	343.2	1,519.7	1,519.7	1,519.7	784	784	784	13,888	13,888	13,888
SOR	21.3	193.5	774.2	94.5	193.5	774.2	84	96	96	372	384	384
stfrg	0.2	4.6	6.1	0.7	4.6	6.1	42	64	48	186	256	192
tomcatv	57.1	1,040	8,323.2	252.9	2,072.6	8,323.2	112	128	128	496	1,024	1,024
swim256	57.6	914.4	3,649.1	229	914.4	3,649.1	128	142	142	464	478	478

Table 4: % improvement over `base` (message vectorized) in communication volume and number of data messages.

PROGRAM	% improvement in communication volume						% improvement in number of data messages					
	Number of Processors = 8			Number of Processors = 32			Number of Processors = 8			Number of Processors = 32		
	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC	BLOCK	B-CYC	CYCLIC
Jacobi	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%
2D hydro	50%	50%	50%	50%	50%	50%	50%	51%	50%	50%	50%	51%
ADI	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%
vpenta	1%	1%	1%	1%	1%	1%	79%	78%	78%	78%	78%	78%
SOR	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%
stfrg	71%	64%	0%	71%	64%	0%	83%	74%	0%	81%	73%	0%
tomcatv	74%	74%	75%	75%	75%	75%	74%	74%	75%	75%	75%	74%
swim256	37%	41%	43%	41%	41%	43%	44%	42%	42%	43%	41%	43%
average	29%	29%	21%	30%	29%	21%	41%	40%	31%	41%	40%	31%

Table 5: % improvement in number of synchronization messages obtained by our approach over the message vectorized (`base`) and globally optimized (`C-Opt`) versions.

PROGRAM	Number of processors = 8						Number of Processors = 32					
	BLOCK		B-CYC		CYCLIC		BLOCK		B-CYC		CYCLIC	
	Base	C-Opt	Base	C-Opt	Base	C-Opt	Base	C-Opt	Base	C-Opt	Base	C-Opt
Jacobi	100%	0%	100%	0%	100%	0%	100%	0%	100%	0%	100%	0%
2D hydro	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
ADI	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
vpenta	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
SOR	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
stfrg	83%	0%	75%	0%	83%	0%	83%	0%	75%	0%	83%	0%
tomcatv	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
swim256	94%	89%	93%	89%	93%	89%	93%	88%	93%	88%	93%	88%
average	97%	74%	96%	74%	97%	74%	97%	74%	96%	74%	98%	74%

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